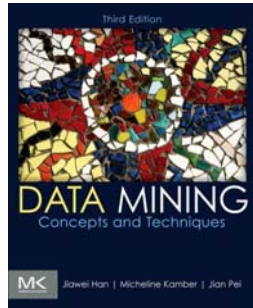


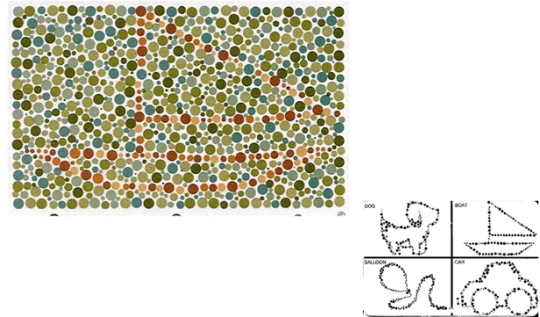
Data Mining



CSC 240/440 by Prof. Jiebo Luo

1

Vision Test?

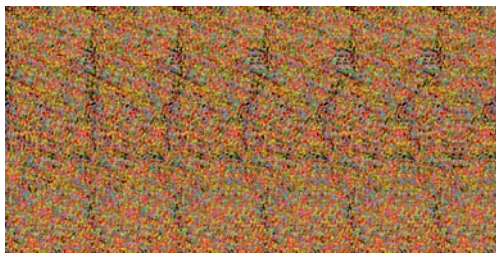


August 29, 2019

CSC 240/440 by Prof. Jiebo Luo

2

Vision Test, or Data Mining?



August 29, 2019

CSC 240/440 by Prof. Jiebo Luo

3

Why taking data mining?

- You
 - You background (UG/MS/PhD)
 - Your expectation
- My expectation:
 - You have taken MTH 161/165, CSC 171/172
 - Preferably you have taken CSC 242 or 262
 - You love to work with data (or you think)

CSC 240/440 by Prof. Jiebo Luo

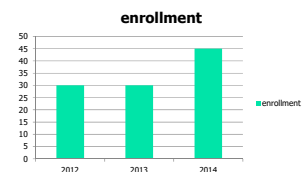
4

Why taking data mining?



5

Capping the class size



What happened since 2015? 5

CSC 240/440 by Prof. Jiebo Luo

6

Quote of the Day



CSC 40/440 by Prof. Jiebo Luo

7

What is data mining?

The *computerized* (sometimes iterative and interactive) process of discovering valid, novel, useful, and understandable patterns or models in

Massive databases

CSC 240/440 by Prof. Jiebo Luo

8

Big Data



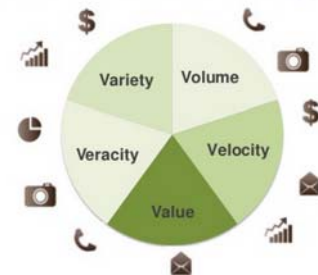
August 29, 2019

9

9

The 5 Vs of Big Data

To get a better understanding of what Big Data is, it is often described using Five Vs:



August 29, 2019

10

10

What is data mining?

- Valid: generalize to the future
- Novel: what we don't know
- Useful: be able to take some action
- Understandable: leading to insight
- Iterative: takes multiple passes
- Interactive: *human* in the loop

CSC 240/440 by Prof. Jiebo Luo

11

Data mining goals

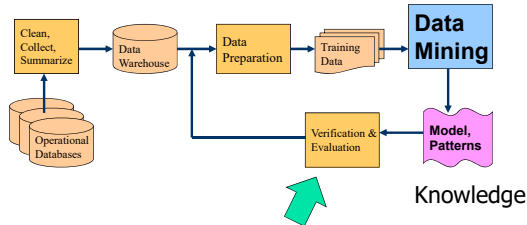
- Prediction
 - Predict the value of a specific attribute based on the values of other attributes, e.g., medical diagnosis
 - "Opaque": ->What?
 - Approaches: classification, regression, outlier detection
- Description
 - Derive patterns (correlation, trends, trajectories) that summarize the underlying relationship between data, e.g., trending in twitter
 - "Transparent": ->Why?
 - Approaches: clustering, association rules, pattern discovery

CSC 240/440 by Prof. Jiebo Luo

12

KDD Process

Knowledge Discovery from Data



CSC 240/440 by Prof. Jiebo Luo

13

Data mining process

- Understand the application *domain*
 - Prior knowledge, user goals
- Create a target dataset
 - Select data, focus on subsets (*relevant to the task*)
- Data cleaning and transformation
 - Remove noise, outliers, missing values
 - Select features, reduce dimensions



CSC 240/440 by Prof. Jiebo Luo

14

Data mining process

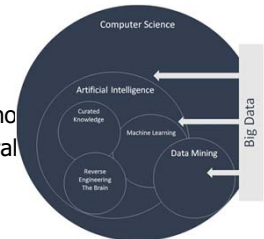
- Apply data mining algorithms
 - Associations, sequences, classification, clustering, etc.
- Interpret, evaluate and visualize patterns
 - What's new and interesting?
 - Iterate if needed
- Manage discovered knowledge
 - Close the loop!

CSC 240/440 by Prof. Jiebo Luo

15

Related fields

- Big data
- Data science
- Machine learning
- Statistics
- Databases and data warehouse
- High performance and parallel computing
- Visualization

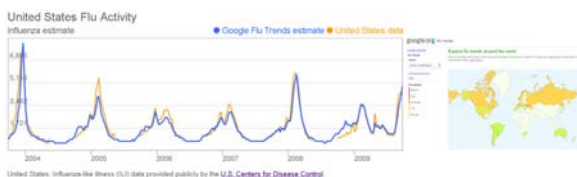


CSC 240/440 by Prof. Jiebo Luo

16

Google Flu Trends

- Conjecture: When people get sick, some of them google the word "flu".
- Statistical analysis of the *crowd* allows the location and severity of the disease to be monitored, up to 2 weeks earlier than CDC using data from hospitals (Ginsberg et al., *Nature*, 2009).



17

CSC 240/440 by Prof. Jiebo Luo

17

Information Networks Are Everywhere



18

Heterogeneous Networks

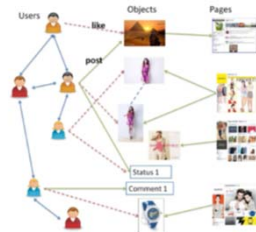


Figure 2: A heterogeneous network model for social media with "like" function, using Facebook as an example. A blue bidirectional arrow is a friendship link. A red dashed arrow is a like action, while a green arrow denotes a post action, annotated with the time stamp when it was posted. The dashed blue line shows that the two photos are visually similar.

19

19

The Wisdom of Crowds

- Independent input from non-experts can sometimes accurately predict outcomes better than experts



- Examples:

- In a 1906 fair, the mean of 787 guesses at the weight of a bull was within a pound
- The crowd ("Ask the audience" in *Who Wants to Be a Millionaire?*) select the correct answer more often than the experts ("Phone a friend")
- Maze navigation improves with crowd input
- The stock market is a massive "Stupidity of crowds"



20

CSC 240/440 by Prof. Jiebo Luo

20

A Social Picture is Worth 1000 Votes

Ge Ma, Jiebo Luo, ICME 2013

- Image sentiment

- Positive vs. negative campaigns
- Positive vs. negative social media
- Flattering vs. unflattering expressions



21

2012: Calling the Swing States

- Acts like a prism to reveal the spectrum of opinions
- Competitive Vector Autoregressive Model

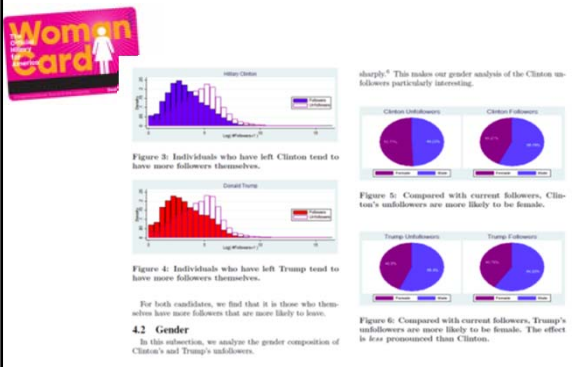


Table 2: Prediction for the swing states on Election Day

State	Official		AR		Flick-AR		VAR		CVAR	
	Obama	Romney	Obama	Romney	Obama	Romney	Obama	Romney	Obama	Romney
CO	0.5239	0.4761	0.5073	0.4927	0.5076	0.4924	0.5059	0.4941	0.5049	0.4951
FL	0.5044	0.4956	0.4936	0.5064	0.4928	0.5072	0.4917	0.5083	0.5018	0.4982
IA	0.5287	0.4713	0.5139	0.4861	0.5132	0.4868	0.5153	0.4847	0.5291	0.4709
NC	0.4891	0.5109	0.4851	0.5149	0.4845	0.5155	0.4866	0.5134	0.4488	0.5512
NV	0.5335	0.4665	0.5144	0.4856	0.5144	0.4856	0.5165	0.4835	0.5362	0.4638
OH	0.5098	0.4902	0.5155	0.4845	0.5150	0.4850	0.5105	0.4895	0.5120	0.4880
VA	0.5157	0.4843	0.5022	0.4978	0.5019	0.4981	0.5013	0.4987	0.5185	0.4815
WI	0.5339	0.4661	0.5218	0.4782	0.5219	0.4781	0.5285	0.4715	0.5460	0.4540

22

Fine-Grained Analysis of the 2016 Election



23

America Tweets China: Analysis of State and Individual Characteristics Regarding Attitudes towards China

Main Contributors: Yu Wang and Jiebo Luo, IEEE Big Data Conference, 2015

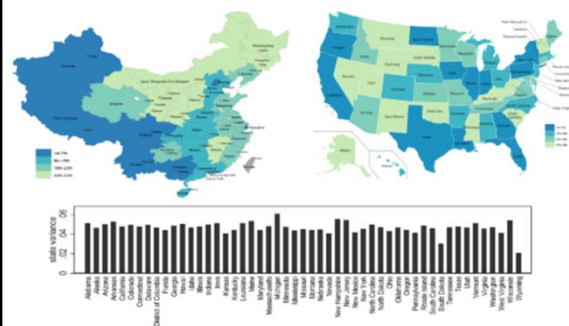


Figure 5: Friendliness and Variance. The top figure reports the friendliness index for each state. The bottom figure reports the variance index for each state.

24



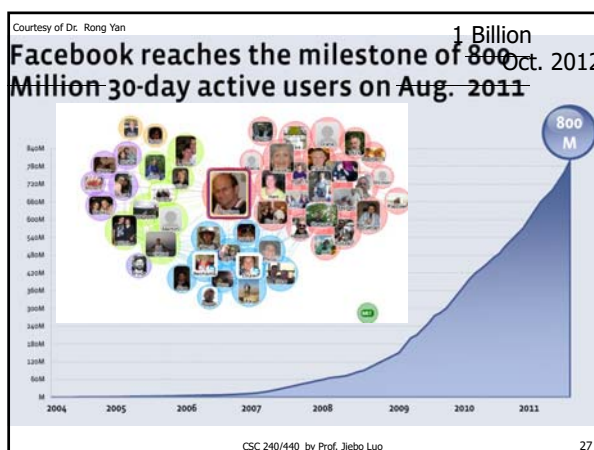
25

Twitter Health
 Adam Sadilek, Henry Kautz, and Vincent Silenzio, Predicting Disease Transmission from Geo-Tagged Micro-Blog Data, AAAI 2012

- You can explore health patterns with the web application at [GermTracker](#)

26

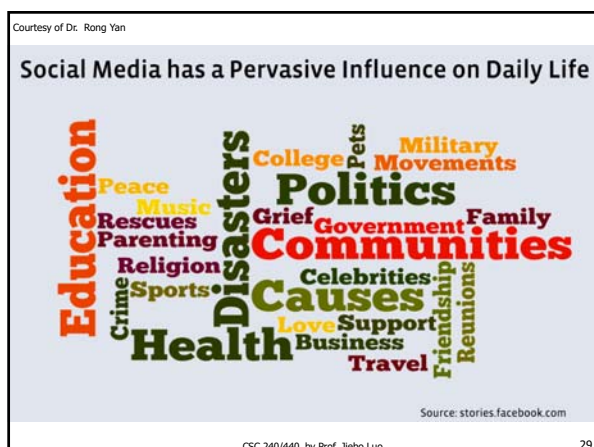
26



27



28



29

Courtesy of Dr. Rong Yan

Mining Large-Scale Social Multi-Media

Understand users better and improve their social experience?

Connection	<ul style="list-style-type: none"> Who should be suggested to you as new friends How to measure and discover online communities
Communicate	<ul style="list-style-type: none"> What are users talking about in their news feed How to analyze user rating and reputation User sentiment and emotion in blogosphere
Influence	<ul style="list-style-type: none"> How does the memes influence user behavior What are your viral marketing strategies
Identity	<ul style="list-style-type: none"> What kinds of ads are relevant to a user Can we suggest face tags for each photo How to recommend the best images / videos

People Analytics

30

30

Courtesy of Dr. Rong Yan

The Pursuit of Happiness

- **Status update** could be an measure of happiness
 1. Representative (unbiased sampled population)
 2. Unobtrusive (based on naturalistic behavior)
 3. Computational (no human raters)
 4. Correlated (co-vary with other valid measures, e.g., non-naturalistic self-reports)
 5. Efficient (process millions of updates per day)

CSC 240/440 by Prof. Jiebo Luo 31

31

Courtesy of Dr. Rong Yan

Facebook Gross National Happiness

- Estimate each country's average happiness based on status updates [http://apps.facebook.com/gnh_index]

August 29, 2019 CSC 240/440 by Prof. Jiebo Luo 32

32

Facebook Mood Manipulation

- Facebook has been experimenting on users. A [new paper in the Proceedings of the National Academy of Sciences \(PNAS\)](#) reveals that Facebook intentionally manipulated the news feeds of almost 700,000 users in order to study "emotional contagion through social networks."
- Army Research Office + Cornell University
- Over the course of the study, it appears, the social network made some of us happier or sadder than we would otherwise have been. Now it's made all of us more mistrustful!

33

33

Big data

IBM's big data portfolio of products for the big data platform

Why IBM for big data

Big data represents a new era in data exploration and utilization. IBM is uniquely positioned to help clients design, develop and execute a big data strategy that will enhance and complement existing systems and processes.

Featured Products

- InfoSphere Streams**
Enables continuous analysis of massive volumes of streaming data with sub-second response times.
- InfoSphere BigInsights**
An enterprise-ready, Apache Hadoop-based solution for managing and analyzing massive volumes of structured and unstructured data.
- InfoSphere Data Explorer**
Discovery and navigation software that provides real-time access and fusion of big data with rich and varied data from enterprise applications for greater insight and ROI.
- IBM PureData powered by Heterogeneous**
Simplifies and optimizes performance of data services for analytic applications, enabling easy complex algorithms to run on unrelated data.
- QDB with BLU Acceleration**
Advanced, innovative capabilities to accelerate analytic workflows for databases and data warehouses.
- IBM Smart Analytics System**
Provides a comprehensive portfolio of data management, hardware, software, & services capabilities that modularly delivers a wide assortment of business-changing analytics.
- InfoSphere Master Data Management**
Creates trusted views of your master data for improving your applications and business processes.
- InfoSphere Information Server**
Understand, cleanse, transform and deliver trusted information to your critical business initiatives, integrating big data into the rest of your IT systems.

Contact IBM
Considering a purchase?
• Chat now
• Email IBM
• Request a quote
• Call us at 800.800.8070
Priority code: 100000000

See what's new!
• InfoSphere BigInsights v2.1
• InfoSphere Streams v2.1
• InfoSphere Information Server v2.1
• InfoSphere Master Data Management v2.1

IBM's big data platform
IBM has developed a comprehensive, integrated and scalable strategy for big data that allows you to address the full spectrum of big data business challenges.

August 29, 2019 Data Mining: Concepts and Techniques

34

August 29, 2019 Data Mining: Concepts and Techniques 35

35

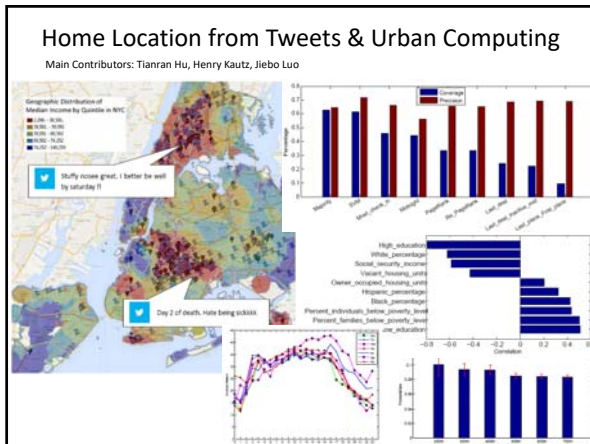
Credit Card Fraud Detection

Trouble with rule-based approaches

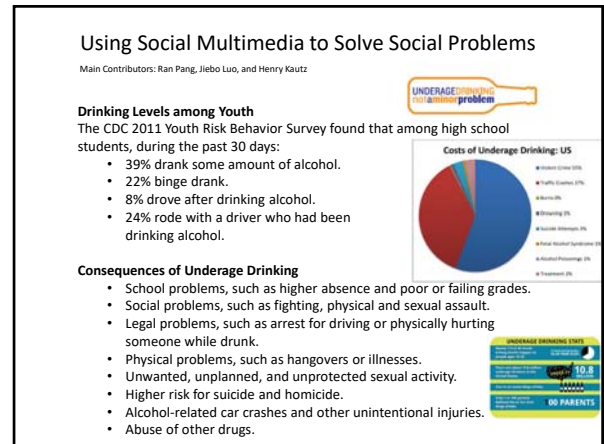
- New fraud schemes constantly "invented" by criminals
- Expensive to build (knowledge intensive)
- Difficult to maintain: frequently rules have to be updated

August 29, 2019 Data Mining: Concepts and Techniques 36

36



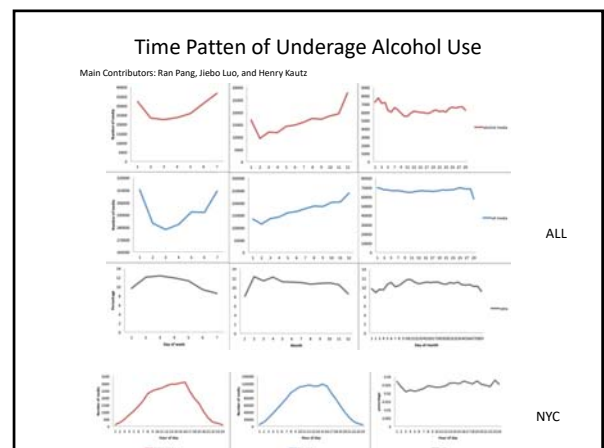
37



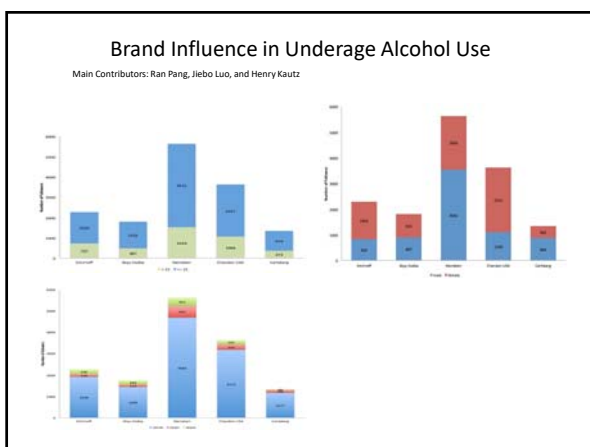
38



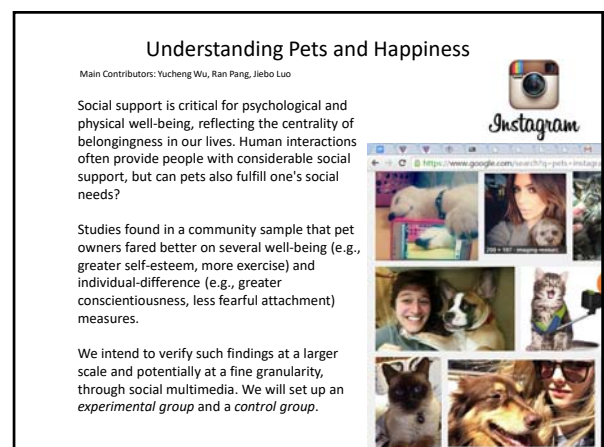
39



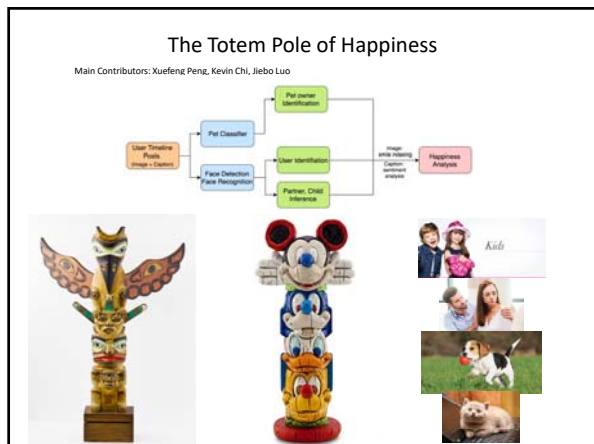
40



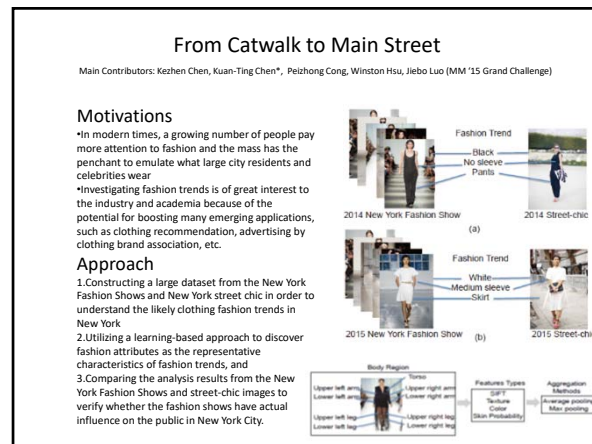
41



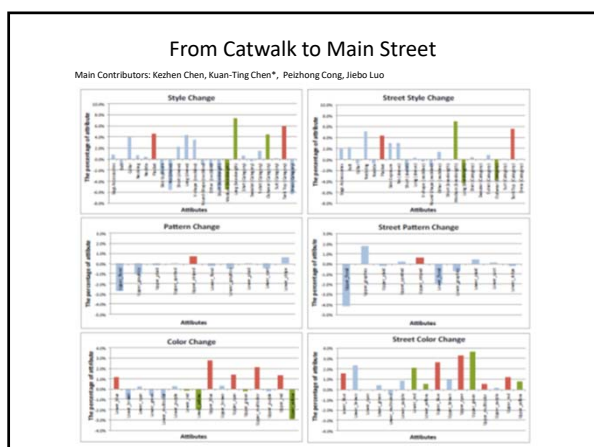
42



43



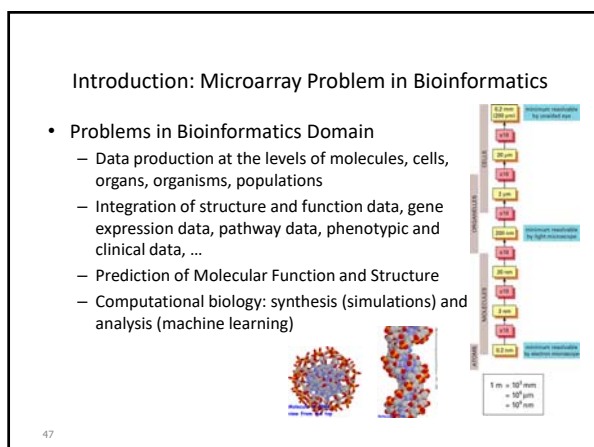
44



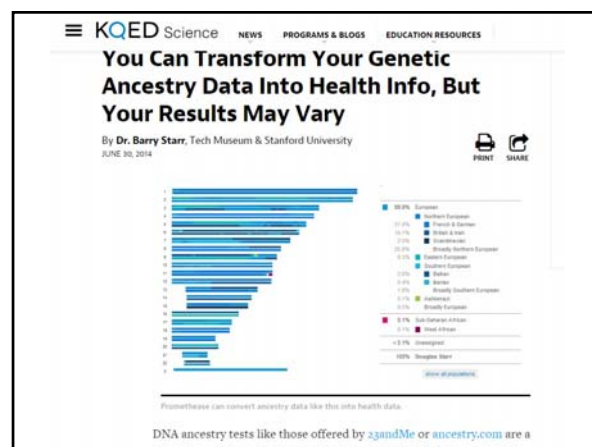
45



46



47



48

Data Mining for Various Applications

- Mining data streams: Tilted time-window, micro-clustering, biased sampling & ensemble
- Mining spatial and multimedia data: Progressive deepening, co-location, invariant analysis
- Mining moving objects, trajectories, RFID/sensor networks
 - Sub-trajectory partitioning and regrouping
- Mining text and Web: PageRank to Web community discovery, opinion & usage mining, veracity analysis, ...
- Mining software bugs, system performance data
- Mining biological data
- Privacy-preserving data mining
- Invisible data mining

49

Textbook

- Required: "Data Mining: Concepts and Techniques", 3/E, by Jiawei Han, Micheline Kamber and Jian Pei. All the homework assignments are from this book.
- Optional: "Mining of Massive Datasets", 2/E, Jure Leskovec, Anand Rajaraman, Jeffery David Ullman. More advanced.
- Course webpage: through my URCS page, or Blackboard (up to date)

CSC 240/440 by Prof. Jiebo Luo

50

Syllabus/Topics (2019)

- Overview and Introduction (notes, Chap. 1)
 - Getting to Know Your Data (Chap. 2)
 - Data Preprocessing (Chap. 3)
 - Linear Algebra, Statistics Review (notes)
 - Pattern Recognition Concepts (notes, Duda & Hart)
 - Mining Frequent Patterns (Chap. 6)
 - Association and Correlation (Chap. 6)
 - Advanced Pattern Mining (Chap. 7)
 - Classification (Chap. 8/9*)
 - Cluster Analysis (Chap. 10/11*)
 - Outlier detection (Chap. 12)
 - Advanced Topics: Social Media Mining (Special Lecture)
 - Advanced Topics: Bioinformatics (Guest Lecture: Martin Zand/Tim Dye)
 - Advanced Topics: Network Mining* (Guest Lecture*: Gourab Ghoshal)
 - Trends and Research Frontiers (Chap. 13, notes)
- * time permitting

CSC 240/440 by Prof. Jiebo Luo

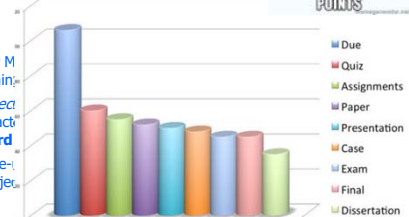
51

Grading

- Homework 25% (5 x 5)
- Mid-term 30% (close book)
- Project I 10% (same assignment to)
- Project II 30% (including presentation)
- Attendance/effort 5%

COURSE POLICY:

- **Course Prerequisites:** M Python/Java programming
- **Homeworks and Project points** will be subtracted submit in **Blackboard**
- **Exam policy:** No make-up, a valid reason, the project



54

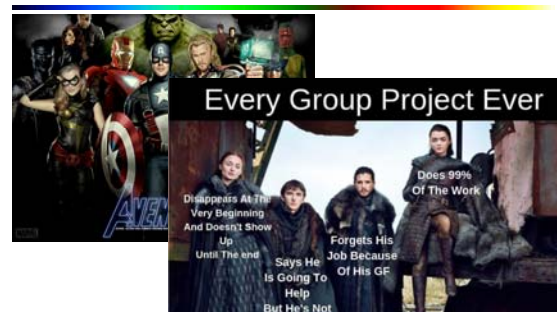
Course Project

- Something 'NEW' is expected
 - New problem, new method
 - Old problem, new method
 - Old method, new problem
 - New comparative study
 - Setting a high target: a conference paper
 - 2-person teams (both must present!)
 - Project proposal
 - Problem statement & related work
 - Data source
 - Software tools and/or programming language(s)
 - Main references
 - Project presentation (1 in 2 selected per *interestingness*)
- Who needs papers? **You** do!

CSC 240/440 by Prof. Jiebo Luo

55

Team Project?



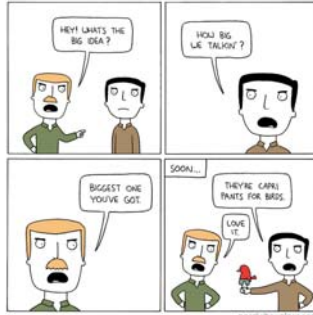
August 29, 2019

Data Mining: Concepts and Techniques

56

More on course projects

- Your own idea
- Suggested topics



August 29, 2019

Data Mining: Concepts and Techniques

57

57

What Can You Expect to Accomplish?

- General knowledge of the field
- Major concepts/approaches
- Exposure to extensive real-world problem-solving examples
- Hands-on experience with solving a real-world data mining problem
- Appreciation of data and data mining
- Awareness of the field: What's hot today? Where is DM going?
- Inspiration – Where can you use DM?

CSC 240/440 by Prof. Jiebo Luo

58

What You **CANNOT** Expect

- Data science is for *everyone*
- An easy course with *light* materials [to get B or better]
- **Survival** without *programming* skills (CSC 171/172)
- **Success** without general knowledge of AI (e.g. CSC 242), math (CSC 161/164) and probability (e.g. CSC 262)
- Doing well while skipping classes and spending little time on textbook
- Getting good grades on assignments by working only the night before the due date
- Giving practice questions for midterm exam (would your future boss do that? He does not even know!)
- A lot of hand-holding as in typical undergraduate courses (This is an advanced course; I will NOT spend one lecture going through one equation: I WILL help you though)
- Getting an "A" easily (>30% students DID get A/A-)
- Whining about any of the above in the course evaluation (No, undergraduates do NOT have any disadvantages)

CSC 240/440 by Prof. Jiebo Luo

59

Teaching Philosophy

- I will teach a 400-level course
- Not 200-level (grow up, or wait until grown up) 6
- For those taking 240: if you meet the same requirements, you get a bonus of 5%; if not quite, you get a prorated bonus.
- Certainly not 100-level
- Can anyone explain the differences?
- How do you know if you are ready for it?
 - Prerequisites: B or better
 - Freshman/sophomore (what's the hurry?), non-CS: placement test
 - "What if I'm not ready but required to take it now"
- Are you up to the challenge?



CSC 240/440 by Prof. Jiebo Luo

60

Testimonials

- "It gives students the freedom to study whatever sub-topics they are most interested in. I spent most of my time studying Bioinformatics, but other students spent time studying card games. There is a high level of freedom."

August 29, 2019

Data Mining: Concepts and Techniques

62

62

Testimonials



Ulrik Soderstrom '16, '17 (MS) is one of the first Rochester students to graduate with a BA in data science.

Soderstrom is interested in the intersection of data science and environmental sustainability.

- **What's your favorite class that you've taken at the University of Rochester?**
- "Last semester, Data Mining with Jiebo Luo was an excellent course. I think he does a good job of giving students a range of skills that go just in-depth enough and then allowing them a creative space for an open-ended project at the end. I worked on a data mining project that determined the effects of industry-specific events—such as the impact of the presidential debates—on stock differences of companies like General Motors or Exxon. That's really rare in a lot of academics to get free rein like that and be responsible for your project."

August 29, 2019

Data Mining: Concepts and Techniques

63

63

A (sometimes) dreaded question



CSC 240/440 by Prof. Jiebo Luo

64

A sometimes (dreaded) question



CSC 240/440 by Prof. Jiebo Luo

65

Making This Course Work

- Working together
 - We have a very diverse, very large class
 - We will cover a lot of materials
 - Your cooperation: assignments on time, study on your own
 - You are here to learn, not just to get a grade
- Lectures: Powerpoint (occasionally board)
 - You** should take notes
 - You** should read the textbook (*before/after* the lecture)
- Course Schedule
 - Guest lectures
 - Instructor's office hours (T/R 3-4pm)
 - Up to 4 TAs: Tianlang Chen (M/W 3:30-4:30pm), Numair Sani (M/W 1-2pm), Yiming Pan (T/R 2-3pm) in WEGMANS 3504
- Communication
 - Blackboard** (only using emails when necessary)

CSC 240/440 by Prof. Jiebo Luo

66

An Alternative

- Business Analytics (offered by Simon School)
 - Lighter treatment
 - Spreadsheet/R
- Do what's right for YOU, ignore the HYPE



August 29, 2019

Data Mining: Concepts and Techniques

67

Academic Honesty

- You are responsible for knowing the College of Arts, Sciences and Engineering and course policies on academic honesty. Read your syllabus for the academic honesty policy for this course. Talk to me if you have any questions.
- I take violations of academic honesty seriously. Suspected violations will be pursued vigorously following the College's procedures for suspected cases of academic dishonesty (see URL link below)

Some Categories of Academic Dishonesty

- Plagiarism:** Representing someone else's work as your own
- Cheating:** Using unauthorized information or sources for an assignment or exam.
- Assisting others in academic dishonesty**
- Falsifying Information**
- Interfering with others' access to legitimate course materials**

This is not an all-inclusive list. For more information, consult with your instructor, the student handbook and visit the URL below.
<http://www.rochester.edu/College/Honesty>

68

Academic Honesty

Summarized by George Ferguson

- Responsibilities of instructors: <http://www.rochester.edu/college/honesty/instructors.html>
 Include an academic honesty statement on each course syllabus or the course Blackboard page or website about how academic honesty applies to the course, and call attention to the information during at least one class session during the first two weeks of class. [continues...]
- On the question of graduate students being covered by the Academic Honesty policy: <http://www.rochester.edu/college/honesty/returninggraduates.html>
 "Graduate students in Arts, Sciences, and Engineering are now included under the College Academic Honesty Policy. Before, if a graduate student were accused of academic dishonesty, the case would be resolved within their department before going to the Graduate Dean of AS&E. To improve consistency and increase fairness, if a graduate student is suspected of academic dishonesty in their coursework, it will now be resolved through Board on Academic Honesty processes."
- Note also that the excuse of "inexperience" is not permitted for graduate students: <http://www.rochester.edu/college/honesty/instructors.html#suspect>
- On the question of the honor pledge in quizzes, the wording says "all examinations": <http://www.rochester.edu/college/honesty/policy.html#pledge>
 "The following Honor Pledge will be copied and signed by all students on all examinations: 'I affirm that I will not give or receive any unauthorized help on this exam, and that all work will be my own.'"

August 29, 2019

Data Mining: Concepts and Techniques

69

From a Computer Science course

- 2) A model for courses that involve collaborative work in laboratory or problem sets, :
- Academic honesty: www.rochester.edu/college/honesty/
www.rochester.edu/college/CCAS/AdviserHandbook/AcadHonesty.html
Homework/Project collaboration: **You may discuss homework problems with others**, but you must **not retain** written notes from your conversations with other students, or share data via computer files to be used in completing your homework. Your written work must be completed without reference to such notes, with the exception of class and recitation notes, which may be retained in written form. [NOTE: some instructors require students to report the names of those with whom they discussed an assignment.]
- General rule: **When in doubt, cite (algorithm, code, data from online)**

CSC 240/440 by Prof. Jiebo Luo

70

Aptitude/Placement Test

- Absolute basics
 - Freshman/Sophomore, non-CS
 - Quantitative
 - Programming
 - Honesty
 - Meeting deadlines
- Midterm: October 26 or 31
Project Presentation:
11/30, 12/5, 12/7, 12/12



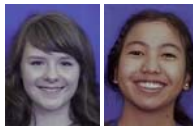
August 29, 2019

Data Mining: Concepts and Techniques

71

Want to do well in 240/440?

- Absolutely no plagiarism
- Follow my instructions (religiously)
 - Meet the prerequisites!
 - Take notes
 - Read the textbook
 - Work on assignments early and independently
 - Make use of office hours
 - Prepare for midterm
 - Put effort in the course project
 - Forget about whining!
 - If you claim you meet the prerequisites
 - If you don't turn in assignments
 - If you skip classes w/o a good reason
 - ...



August 29, 2019

Data Mining: Concepts and Techniques

72